

Tutorial Sheet - 1

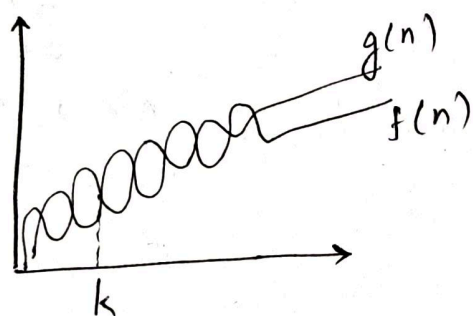
①

Sol 1 :- Asymptotic Notation:- These notations are used to tell the complexity of an algorithm when the i/p is very large.

→ It describes the algorithm efficiency and performance in a meaningful way. It describes the behaviour of time or space complexity for large instance characteristics.

• The asymptotic notation of an algorithm is classified into 5 types :-

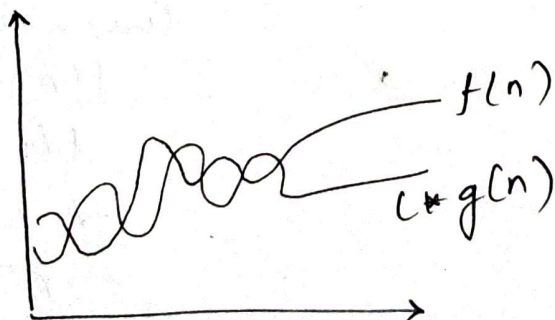
(i) Big Oh notation (O) :- (Asymptotic upper Bound) The function $f(n) = O(g(n))$, if and only if there exist a +ve constant C and k such that $f(n) \leq C * g(n)$ for all $n \geq k$.



$$\begin{aligned} f(n) &= O(g(n)) \\ \text{iff} \\ f(n) &\leq C \cdot g(n) \\ &\forall n \geq n_0, \\ &\text{some constant } C > 0 \end{aligned}$$

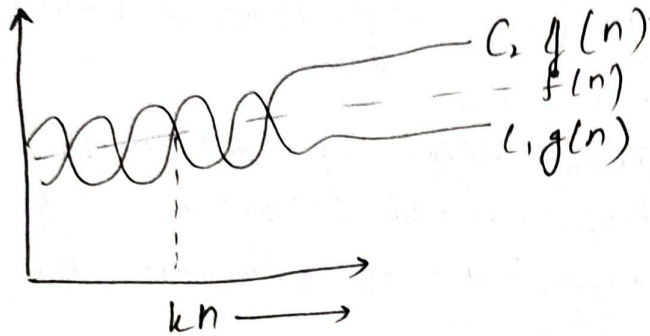
(ii) Big Omega notation (Ω) :- (Asymptotic lower bound).

The function $f(n) = \Omega(g(n))$, iff there exists a +ve constant C and k such that $f(n) \geq C * g(n)$ for all $n, n \geq k$.



$$\begin{aligned} f(n) &= \Omega(g(n)) \text{ iff} \\ f(n) &\geq C \cdot g(n) \\ &\forall n \geq n_0 \text{ \& some} \\ &\text{const } C > 0 \end{aligned}$$

(iii) Big theta notation (θ):- (Asymptotic tight bound):- ②
 The function $f(n) = \theta(g(n))$, iff there exists a +ve constant c_1, c_2 , and k such that $c_1 * g(n) < f(n) < c_2 * g(n)$ for all $n \geq k$.



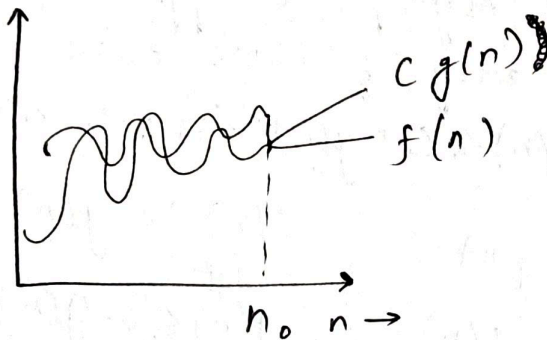
$$f(n) = \theta(g(n))$$

iff

$$c_1 g(n) \leq f(n) \leq c_2 g(n)$$

$$\forall n \geq \max(n_1, n_2)$$

(iv) Small-o notation (o):- o gives us upper bound
 $f(n) = o(g(n))$.



$$f(n) < c g(n)$$

$$\forall n > n_0 \text{ \& \& } \forall c > 0$$

$$n = o(n^2)$$

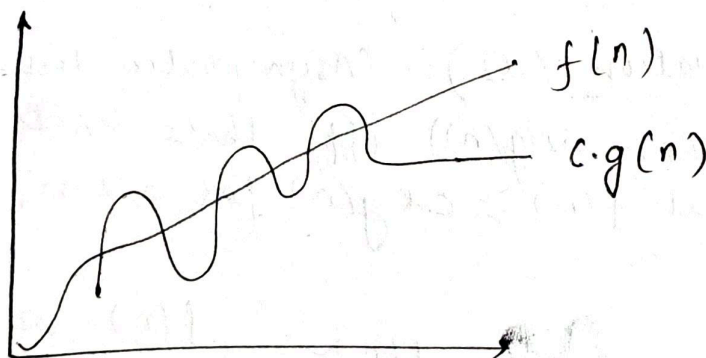
$$n < 1n^2$$

$$2n^2$$

$$0 < n^2$$

$$n < 0.001 n^2 n_0$$

(v) Small-omega notation (ω):-



lower bound

$$f(n) = \omega(g(n))$$

$$f(n) > c \cdot g(n)$$

$$\forall n > n_0 \text{ \& \& } \forall c > 0$$

$$n^2 = \omega(n)$$

Sol 2 :- for (i=1 to n)
 {
 i = i * 2;
 }

(3)

Time complexity for a loop means no. of times the loop has to run.

→ For the loop above, the loop will run for following values of i :-

i	1	2	4	8	16	32	64	...	2^k
value	2^0	2^1	2^2	2^3	2^4	2^5	2^6		2^n

i = 1, 2, 4, 8, 16, 32, ... 2^k this means k times

i.e. $2^k = n$ [Taking log both sides]

$$k \log_2 2 = \log_2 n$$

$$k = \log_2 n \quad [\because \log_a a = 1]$$

$$\therefore \boxed{T.C = O(\log n)}$$

Sol 3. $T(n) = \begin{cases} 3T(n-1), & n > 0 \\ 1 & \end{cases}$

By forward substitution,

$$T(n) = 3T(n-1)$$

$$T(0) = 1$$

$$\begin{aligned} T(1) &= 3T(1-1) \\ &= 3T(0) \\ &= 3 \end{aligned}$$

$$\begin{aligned} T(2) &= 3T(2-1) \\ &= 3T(1) \\ &= 3 \times 3 = 3^2 \end{aligned}$$

$$\begin{aligned} T(3) &= 3T(3-1) \\ &= 3T(2) \\ &= 3 \times 3^2 = 3^3 \end{aligned} \quad \dots \dots T(n) = 3^n$$

$$\therefore \boxed{TC = O(3^n)}$$

(4)

Sol 4.) $T(n) = \begin{cases} 2T(n-1) - 1, & n > 0 \\ 1 & \end{cases}$

By forward substitution

$$T(0) = 1$$

$$\begin{aligned} T(1) &= 2T(1-1) - 1 \\ &= 2T(0) - 1 = 2T(0) - 1 \\ &= 2 \times 1 - 1 = (2 - 1) \end{aligned}$$

$$\begin{aligned} T(2) &= 2T(2-1) - 1 \\ &= 2T(1) - 1 \\ &= 2(2-1) - 1 = 2^2 - 2^1 - 1 \end{aligned}$$

$$\begin{aligned} T(3) &= 2T(3-1) - 1 \\ &= 2T(2) - 1 \\ &= 2(2^2 - 2^1 - 1) - 1 \\ &= 2^3 - 2^2 - 2^1 - 1. \end{aligned}$$

⋮

$$T(n) = 2^n - 2^{n-1} - 2^{n-2} - \dots - 2^2 - 2^1 - 2^0$$

$$\begin{aligned} &= 2^n - (2^n - 1) \\ &= 2^n - 2^n + 1 = 1 \end{aligned}$$

$$\therefore \boxed{T(n) = O(1)}$$

Sol 5.)

```
int i = 1, s = 1
while (s <= n)
{
    i++;
    s = s + i;
    printf("%d #");
}
```

$$S_i = S_{i-1} + i$$

(5)

The value of 'i' increases by one for each value contained in 's' at the i^{th} iteration is the sum of the first 'i' +ve integers. If k is the total no. of iterations taken by any program then while loop terminates if: $1+2+3+\dots+k$

$$= [k(k+1)/2] > n$$

$$\text{So } k = O(\sqrt{n})$$

$$\therefore \boxed{T.C = O(\sqrt{n})}$$

Sol 6.)

void function (int n)

{

int i, count=0;

for (i=1; i<=n; i++)

count++;

}

} $O(n)$

$$\boxed{T.C = O(n)}$$

Sol 7.)

void function (int n)

{

int i, j, k, count=0;

for (i=n/2; i<=n; i++) $\rightarrow O(n)$

for (j=1; j<=n; j=j*2) $\rightarrow O(\log n)$

for (k=1; k<=n; k=k*2) $\rightarrow O(\log n)$

count++;

}

$$T.C = n * \log n * \log n$$

$$\boxed{T.C = O(n \log^2 n)}$$

Sol 8.)

function (int n)

{

if (n==1)

return;

for (i=1 to n) $\rightarrow O(n)$

{

for (j=1 to n) $\rightarrow O(n)$

{

printf("%*");

}

}

function (n-3);

}

Time complexity = $O(n^2)$ Sol 9.)

void function (int n)

{

for (i=1 to n) { $\rightarrow O(n)$ for (j=1; j<=n; j=j+1) $\rightarrow O(n)$

printf("%*");

}

}

$$T.C = O(n) * O(n) = \boxed{\underline{\underline{O(n^2)}}}$$

Sol 10.) n^k is $O(c^n)$ ans

$$\underline{\underline{n^k = O(c^n)}}$$